Notes: In the sequel, it is advisable to use the command

randn('seed',0)

before generating the data sets, in order to initialize the Gaussian random number generator to 0 (or any other fixed number). This is important for the reproducibility of the results.

Gaussian generator: Generate N l-dimensional vectors from a Gaussian distribution with mean m and covariance matrix S, using the mvnrnd MATLAB function.

## Solution

Just type

mvnrnd(m,S,N)

1.

- **a.** Generate a data set  $X_1$  of N=1,000 two-dimensional vectors that stem from three equiprobable classes modeled by normal distributions with mean vectors  $m_1 = [1, 1]^T$ ,  $m_2 = [12, 8]^T$ ,  $m_3 = [16, 1]^T$  and covariance matrices  $S_1 = S_2 = S_3 = 4I$ , where I is the  $2 \times 2$  identity matrix.
- **b.** Apply the Bayesian, the Euclidean, and the Mahalanobis classifiers on  $X_1$ .
- Compute the classification error for each classifier.

2.

- a. Generate a data set  $X_2$  of N = 1,000 two-dimensional vectors that stem from three equiprobable classes modeled by normal distributions with mean vectors  $m_1 = [1, 1]^T$ ,  $m_2 = [14, 7]^T$ ,  $m_3 = [16, 1]^T$  and covariance matrices  $S_1 = S_2 = S_3 = \begin{bmatrix} 5 & 3 \\ 3 & 4 \end{bmatrix}$ .
  - (b)-(c) Repeat steps b) and (c) of experiment 2.2, for  $X_2$ .

3.

- **a.** Generate a data set  $X_3$  of N=1,000 two-dimensional vectors that stem from three equiprobable classes modeled by normal distributions with mean vectors  $m_1 = [1, 1]^T$ ,  $m_2 = [8, 6]^T$ ,  $m_3 = [13, 1]^T$  and covariance matrices  $S_1 = S_2 = S_3 = 6I$ , where I is the  $2 \times 2$  identity matrix.
  - (b)-(c) Repeat (b) and (c) from experiment 2.2, for  $X_3$ .

4.

- a. Generate a data set  $X_4$  of N=1,000 two-dimensional vectors that stem from three equiprobable classes modeled by normal distributions with mean vectors  $m_1 = [1,1]^T$ ,  $m_2 = [10,5]^T$ ,  $m_3 = [11,1]^T$  and covariance matrices  $S_1 = S_2 = S_3 = \begin{bmatrix} 7 & 4 \\ 4 & 5 \end{bmatrix}$ .
  - (b)-(c) Repeat steps (b) and (c) of experiment 2.2, for  $X_4$ .

5.

- a. Generate two data sets  $X_5$  and  $X_5'$  of N=1,000 two-dimensional vectors each that stem from three classes modeled by normal distributions with mean vectors  $m_1 = [1, 1]^T, m_2 = [4, 4]^T, m_3 = [8, 1]^T$  and covariance matrices  $S_1 = S_2 = S_3 = 2I$ . In the generation of  $X_5$ , the classes are assumed to be equiprobable, while in the generation of  $X_5'$ , the *a priori* probabilities of the classes are given by the vector  $P = [0.8, 0.1, 0.1]^T$ .
- **b.** Apply the Bayesian and the Euclidean classifiers on both  $X_5$  and  $X_5'$ .
- c. Compute the classification error for each classifier for both data sets and draw your conclusions.

6.

Consider the data set  $X_3$  (from experiment (2.4)). Using the same settings, generate a data set Z, where the class from which a data vector stems is known. Apply the k nearest neighbor classifier on  $X_3$  for k = 1 and k = 11 using Z as the training set and draw your conclusions.